



## DIGITAL DISCRIMINATION IN THE DIGITAL AGE: ADDRESSING ALGORITHMIC BIAS IN CHILD WELFARE DECISION-MAKING SYSTEMS

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### ABSTRACT

*The integration of algorithmic decision-making into child welfare systems has introduced both unprecedented opportunities and significant risks in the digital age. While predictive analytics and machine learning promise greater objectivity and efficiency in identifying children at risk, these technologies also risk perpetuating and amplifying longstanding social inequalities. This article critically examines the evolution and implementation of algorithmic tools in child welfare, with a particular focus on the sources and evidence of algorithmic bias, as exemplified by the Allegheny County Family Screening Tool. It explores the constitutional and legal challenges posed by these systems, including due process and equal protection concerns arising from opaque and potentially discriminatory decision-making processes. The article argues for a comprehensive regulatory framework encompassing algorithmic accountability, enhanced procedural protections, robust data governance, and ongoing oversight. Ultimately, it calls for participatory, transparent, and ethically grounded approaches to ensure that digital innovation in child welfare serves the cause of justice, rather than reinforcing digital discrimination.*

**Keywords:** Algorithmic Decision-Making, Allegheny County Family, Screening Tool.

### INTRODUCTION

In 2016, Allegheny County, Pennsylvania, attained prominence as the first jurisdiction in the United States to adopt predictive analytics for screening child abuse hotline calls. When making life-changing decisions about when to investigate households for possible child abuse, the Allegheny Family Screening Tool promises to provide consistency and objectivity. However, after a few years of use, troubling trends surfaced: the algorithm continuously

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identified Black families more frequently than white families in comparable situations, so sustaining the very racial inequities it was intended to eradicate. Instead of addressing human bias, this technology has amplified and mechanised discrimination on a previously unseen scale.

To decide which families should be investigated, which children should be taken from their homes, and how best to distribute resources, child welfare organisations across the US are depending more and more on algorithmic decision-making tools. In order to produce risk ratings that inform caseworker choices, these predictive risk assessment algorithms examine extensive information, such as criminal histories, public benefit utilisation, mental health histories, and neighbourhood variables. A system where technology encourages rather than corrects discriminatory practices is created when these algorithms, however, frequently reinforce and magnify preexisting racial, economic, and social prejudices found in historical child welfare statistics.

The exact proportion of Allegheny County's child welfare decisions that currently use computational methods is not disclosed to the public. The ramifications go much beyond differences in statistics. Fundamental constitutional rights, such as the right to equal treatment under the law, due process protections, and family integrity, are impacted by algorithmic determinations. Technology becomes a tool for systemic discrimination when biased algorithms determine whether families receive supportive services rather than punitive investigations or whether children are taken away from their parents. Furthermore, because families cannot often comprehend, contest, or reverse the automated decisions that ruin their lives, the opacity of these algorithmic systems presents significant due process issues.

The current use of predictive analytics tools without proper bias detection, transparency requirements, and due process safeguards violates constitutional protections and sustains systemic discrimination, even though these tools promise to improve child welfare outcomes through data-driven decision-making. This article makes the case that comprehensive legal frameworks that require mandatory bias testing, transparency requirements, and procedural safeguards that strike a balance between technological innovation and fundamental civil rights protections are desperately needed to regulate algorithmic decision-making in child welfare systems.

There are four sections to this article. The development and present use of algorithmic tools in child welfare systems are examined in Part I. With an emphasis on the Allegheny County case study, Part II examines the causes and proof of algorithmic prejudice. The constitutional and legal issues raised by these systems, such as violations of equal protection and due process, are examined in Part III. A thorough regulatory structure comprising algorithmic accountability measures, improved procedural protections, and continuous oversight mechanisms is proposed in Part IV.

## **EVOLUTION AND IMPLEMENTATION OF ALGORITHMIC TOOLS IN CHILD WELFARE**

**Historical Context: From Human Judgment to Machine Learning:** Child protection organisations have been battling the subjectivity and inconsistent nature of human decision-making for decades. Social workers were often forced to make life-changing decisions on resource allocation, family separation, and child safety while working under extreme pressure and with insufficient information. Analogue risk assessment tools were the first attempt to standardise these procedures, but they were also constrained by the prejudices and expertise of their designers.<sup>12</sup>

This scenario changed in the late 20th and early 21st centuries with the advent of digital data and machine learning. In an attempt to use large administrative databases to find trends, forecast results, and make more objective judgments, agencies started implementing algorithmic risk assessments and predictive analytics.<sup>3</sup> These tools are promised to increase consistency across cases and jurisdictions, reduce human error and subjective bias, optimise the allocation of scarce resources, and provide data-driven support for complex decisions.

### **Types of Algorithmic Tools in Child Welfare –**

Algorithmic tools now serve multiple functions in child welfare systems:

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<sup>1</sup> Emnet Almedom and others, 'Algorithms and Child Welfare: The Disparate Impact of Family Surveillance in Risk Assessment Technologies' (Berkeley Public Policy Journal, 2 February 2021) <https://bppj.studentorg.berkeley.edu/2021/02/02/algorithms-and-child-welfare-the-disparate-impact-of-family-surveillance-in-risk-assessment-technologies/> accessed 16 June 2025.

<sup>2</sup> Hall SF and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>3</sup> Hall SF and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

**Predictive Risk Modelling (PRM):** These models project the probability of future child abuse or placement in foster care by using past data and risk factors.<sup>45</sup>

**Automated Triage and Prioritisation:** Algorithms assist in screening, classifying, and prioritising cases for inquiry to guarantee that the most urgent cases are given attention first.<sup>6</sup>

**Resource Allocation:** Certain systems employ algorithms to establish foster parent remuneration, assign foster care placements, and assess eligibility for services.<sup>7</sup>

**Outcome Prediction:** Case planning and court decisions are informed by advanced algorithms that can predict outcomes like reunification, placement stability, or return of abuse.<sup>89</sup>

### **The Realities of Implementation -**

The real application of algorithms in child welfare is complicated and contradictory, despite their potential:

**Data Restrictions:** A lot of algorithms depend on administrative data that is either unrepresentative, out-of-date, or incomplete. Important data, like complex case notes or family evaluations, are sometimes only available in unstructured language, which makes it challenging for algorithms to process.

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<sup>4</sup> John Kelly, 'In a New Study, Georgia Professor Explores the Use of Artificial Intelligence in the Child Welfare Field' (The Imprint, 1 December 2023) <https://imprintnews.org/top-stories/in-a-new-study-georgia-professor-explores-the-use-of-artificial-intelligence-in-the-child-welfare-field/256116> accessed 15 June 2025.

<sup>5</sup> Hall SF and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>6</sup> Emnet Almedom and others, 'Algorithms and Child Welfare: The Disparate Impact of Family Surveillance in Risk Assessment Technologies' (Berkeley Public Policy Journal, 2 February 2021) <https://bppj.studentorg.berkeley.edu/2021/02/02/algorithms-and-child-welfare-the-disparate-impact-of-family-surveillance-in-risk-assessment-technologies/> accessed 16 June 2025.

<sup>7</sup> Elizabeth Kennedy, 'Algorithmic Harms in Child Welfare: Uncertainties in Practice, Organization, and Street-Level Decision-Making' (Montreal AI Ethics Institute, 2 March 2023).

<sup>8</sup> S F Hall and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) 148 Children and Youth Services Review.

<sup>9</sup> John Kelly, 'In a New Study, Georgia Professor Explores the Use of Artificial Intelligence in the Child Welfare Field' (The Imprint, 1 December 2023) <https://imprintnews.org/top-stories/in-a-new-study-georgia-professor-explores-the-use-of-artificial-intelligence-in-the-child-welfare-field/256116> accessed 15 June 2025.

**Legal and Organisational Restrictions:** In certain situations, caseworkers may be legally compelled to employ algorithmic results, even if doing so goes against their professional judgement or pragmatic considerations.<sup>10</sup>

**Lack of knowledge,** the majority of frontline employees lack technical or statistical knowledge, which causes them to overrely on or distrust algorithmic advice.<sup>11</sup>

**Emphasis on Legal Risk,** priorities are skewed by algorithms that are more intended to reduce an agency's legal risk than to optimise child welfare results.

**Gaming and Manipulation,** workers occasionally falsify algorithmic inputs (such as by exaggerating risk scores to obtain resources), which erodes systemic confidence.<sup>12</sup>

## SOURCES AND EVIDENCE OF ALGORITHMIC BIAS—THE ALLEGHENY COUNTY CASE STUDY

### Understanding Algorithmic Bias in Child Welfare –

Algorithmic prejudice occurs when automated systems consistently discriminate against particular groups, which frequently mirrors or exacerbates preexisting societal injustices. Bias can appear in child welfare at several levels:

**Data Collection:** Excessive surveillance of low-income and minority households may distort historical data, resulting in their overrepresentation in risk models.<sup>13</sup>

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<sup>10</sup> Elizabeth Kennedy, 'Algorithmic Harms in Child Welfare: Uncertainties in Practice, Organization, and Street-Level Decision-Making' (Montreal AI Ethics Institute, 2 March 2023).

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<sup>13</sup> Emnet Almedom and others, 'Algorithms and Child Welfare: The Disparate Impact of Family Surveillance in Risk Assessment Technologies' (Berkeley Public Policy Journal, 2 February 2021)

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**Feature Selection:** By using variables such as past system involvement, neighbourhood, or use of public benefits as stand-ins for race or poverty, the algorithm's logic may be biased.<sup>1415</sup>

**Feedback Loops:** Algorithmic choices have an impact on subsequent data gathering, resulting in self-reinforcing cycles of community monitoring and intervention.<sup>16</sup>

**Outcome Disparities:** Algorithms can have disparate effects even when they are not specifically meant to bias, flagging some groups disproportionately for review or expulsion.<sup>1718</sup>

### **The Allegheny Family Screening Tool (AFST): A Case Study -**

**The Operation of the AFST:** One of the most extensively researched instances of algorithmic risk assessment in child welfare is Pennsylvania's Allegheny Family Screening Tool (AFST). The AFST creates a risk score (1–20) for every child abuse hotline contact by analysing more than 100 characteristics from public agency data, including past complaints of maltreatment, engagement with the criminal justice system, and use of public services.<sup>19</sup> The purpose of this score is to help screeners decide which instances to look into.<sup>20</sup>

**Evidence of Bias and Disparate Impact:** “Multiple independent evaluations have raised concerns about bias in the AFST,

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<sup>15</sup> Tim Dare and Eileen Gambrill, ‘Ethical Issues in the Use of Predictive Analytics in Child Welfare’ (2019) 8(10) Social Sciences 281 <https://www.mdpi.com/2076-0760/8/10/281> accessed 18 June 2025.

<sup>16</sup> Emnet Almedom and, ‘Algorithms and Child Welfare: The Disparate Impact of Family Surveillance in Risk Assessment Technologies’ (Berkeley Public Policy Journal, 2 February 2021) <https://bppj.studentorg.berkeley.edu/2021/02/02/algorithms-and-child-welfare-the-disparate-impact-of-family-surveillance-in-risk-assessment-technologies/> accessed 16 June 2025.

<sup>17</sup> Emnet Almedom and others, ‘Algorithms and Child Welfare: The Disparate Impact of Family Surveillance in Risk Assessment Technologies’ (Berkeley Public Policy Journal, 2 February 2021) <https://bppj.studentorg.berkeley.edu/2021/02/02/algorithms-and-child-welfare-the-disparate-impact-of-family-surveillance-in-risk-assessment-technologies/> accessed 16 June 2025.

<sup>18</sup> Hall SF and others, ‘A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias’ (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>19</sup> John Kelly, ‘In a New Study, Georgia Professor Explores the Use of Artificial Intelligence in the Child Welfare Field’ (The Imprint, 1 December 2023) <https://imprintnews.org/top-stories/in-a-new-study-georgia-professor-explores-the-use-of-artificial-intelligence-in-the-child-welfare-field/256116> accessed 15 June 2025.

<sup>20</sup> Emnet Almedom and others, ‘Algorithms and Child Welfare: The Disparate Impact of Family Surveillance in Risk Assessment Technologies’ (Berkeley Public Policy Journal, 2 February 2021) <https://bppj.studentorg.berkeley.edu/2021/02/02/algorithms-and-child-welfare-the-disparate-impact-of-family-surveillance-in-risk-assessment-technologies/> accessed 16 June 2025.

**Disproportionate Targeting:** Regardless of real risk, families from minority or low-income backgrounds—who engage with public services more frequently—are more likely to earn high risk scores since the AFST uses data from these systems.

**Privacy and Surveillance:** Poor families of colour have a systemic loss of privacy as a result of the utilisation of extensive data from many state agencies, and they are also more closely monitored than wealthy, white families.

**Feedback Loops:** following a high-risk family's investigation and flagging, additional information is gathered about them, raising the possibility that they could be reported once more in the future.

**Opaque Decision-Making,** families and activists are not well-informed on the risk score's calculation or how to contest it if it appears to be unfair or erroneous.”<sup>2122</sup>

### **Broader Patterns and Other Jurisdictions -**

The issues identified in Allegheny County are echoed in other jurisdictions:

**Wisconsin:** According to an anthropological study, child welfare algorithms frequently depended on less illuminating data and gave legal risk reduction precedence over child welfare results.

**Oregon:** Attempts to "correct" algorithmic injustice in reunification prediction tools have brought attention to the difficulty of striking a balance between individual and collective fairness, as well as the possibility that corrective actions could exacerbate already-existing imbalances.<sup>23</sup>

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<sup>22</sup> Elizabeth Kennedy, 'Algorithmic Harms in Child Welfare: Uncertainties in Practice, Organization, and Street-Level Decision-Making' (Montreal AI Ethics Institute, 2 March 2023) <https://montrealethics.ai/algorithmic-harms-in-child-welfare-uncertainties-in-practice-organization-and-street-level-decision-making/> accessed 18 June 2025.

<sup>23</sup> S F Hall and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) 148 Children and Youth Services Review 107903 <https://www.sciencedirect.com/science/article/abs/pii/S0190740922004133> accessed 18 June 2025.



**Georgia:** Despite being designed to forecast placement in foster care, risk scores frequently reflected systemic biases in the underlying data, according to a four-year study on AI tools in child welfare.<sup>24</sup>

## CONSTITUTIONAL AND LEGAL CHALLENGES—DUE PROCESS AND EQUAL PROTECTION

**Due Process Concerns:** The Fourteenth Amendment of the U.S. Constitution ensures that no one can be denied their property, freedom, or life without first undergoing a fair legal process. This covers parents' fundamental right to care for their children in the context of child welfare.

**Contestability and Opacity:** Transparency is frequently lacking in algorithmic decision-making. Families may not comprehend the reasoning behind the choice or have a meaningful chance to challenge it when they are the targets of interventions based on opaque risk scores. Procedural justice is compromised, and systemic trust may be damaged.<sup>25</sup>

**Delegation of Discretion Historically,** Courts have mandated that decisions involving discretion that impact fundamental rights undergo substantial review. Delegating discretion to an algorithm, particularly one that is proprietary or technically complicated, makes court review challenging and raises significant due process issues.<sup>26</sup>

**Anti-discrimination and Equal Protection:** States are forbidden by the Equal Protection Clause from denying anyone the same legal protections. If policies have a disproportionate effect on protected groups and are not adequately justified, they may breach equal protection even if they are facially neutral.

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**Differing Effects:** Even in the absence of overt discriminatory intent, algorithmic techniques that give low-income or minority families higher risk rankings may be illegal if they lead to disproportionate removals or interventions.<sup>2728</sup>

**Legal Challenges and Precedents:** The number of lawsuits contesting algorithmic bias in public services is increasing. Courts are starting to struggle with how to apply conventional legal ideas to this new technology, and plaintiffs contend that biased algorithms violate both statutory and constitutional safeguards.<sup>2930</sup>

**The Problem of the “Black Box”:** A lot of algorithmic systems are “black boxes”, either because of technological complexity or proprietary safeguards, because of this opacity, it is challenging for impacted parties to seek compensation and for courts to maintain responsibility.<sup>31</sup>

## TOWARD A COMPREHENSIVE REGULATORY AND ETHICAL FRAMEWORK

A strong ethical and regulatory framework is necessary to guarantee justice, accountability, and fairness in algorithmic child welfare systems because of the high stakes and intricate dangers involved.

### Measures of Algorithmic Accountability -

**Frequent Bias Audits:** Regular, independent audits of algorithms are necessary to identify and reduce bias before they are deployed and throughout their continued use. Testing for

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<sup>28</sup> Hall SF and others, ‘A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias’ (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>29</sup> Hall SF and others, ‘A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias’ (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

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differential effects on protected groups and evaluating long-term feedback loops should be included in audits.

**Documentation and Transparency:** The design, data sources, feature selection, and rationale of algorithms must all be documented by agencies and made available to stakeholders.<sup>32 33</sup> Families, advocates, and courts should all be able to comprehend and contest algorithmic choices as part of transparency.

**Evaluations of Impact:** Agencies shall carry out impact evaluations before deployment to weigh the advantages and disadvantages, including the possibility of unequal effects on protected groups.<sup>34</sup>

### **Strengthened Procedural Defences -**

**The Right to Clarification:** Families impacted by algorithmic judgments have to be entitled to an understandable and unambiguous explanation of the decision-making process, the data used, and the procedures for contesting or fixing mistakes.<sup>35</sup>

**Obtaining Legal Counsel:** When algorithmic results are utilised as the foundation for intervention or removal, families should have access to legal advice due to the intricacy of algorithmic systems.<sup>36</sup>

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<sup>35</sup> Hall SF and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>36</sup> Elizabeth Kennedy, 'Algorithmic Harms in Child Welfare: Uncertainties in Practice, Organization, and Street-Level Decision-Making' (Montreal AI Ethics Institute, 2 March 2023) <https://montrealethics.ai/algorithmic-harms-in-child-welfare-uncertainties-in-practice-organization-and-street-level-decision-making/> accessed 14 June 2025.

**A Process for Meaningful Appeals:** Algorithmic decision appeals must include a strong and user-friendly procedure that allows for the presentation of evidence, cross-examination of witnesses, and independent expert evaluation.<sup>37</sup>

### **Ongoing Oversight Mechanisms -**

**Boards of Independent Review:** To supervise the development, application, and effects of algorithmic tools, agencies should set up independent review committees including members from the fields of law, data science, ethics, and child welfare.

**Designing with Participation:** To guarantee that a range of viewpoints are taken into account, stakeholders such as advocates, community organisations, and impacted families should be included in the development and assessment of algorithmic systems.

**Collaboration Across Disciplines:** To effectively address technical, legal, and human concerns, cooperation between the fields of law, computer science, social work, and ethics is necessary.”<sup>38</sup>

### **Consent and Data Governance -**

**Transparency of Data Sources:** The data sources utilised to train and run algorithms, as well as their limits and any biases, must be openly disclosed by agencies.<sup>39</sup>

**Data Rights and Consent:** Families ought to be able to access the information that is being gathered on them, give their consent when necessary, and have any inaccurate information corrected.<sup>40</sup>

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<sup>37</sup> Hall SF and others, ‘A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias’ (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>38</sup> Hall SF and others, ‘A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias’ (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

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**Minimising Data:** Algorithms should only use the information required to accomplish their goals in order to reduce the possibility of misuse and privacy violations.<sup>41</sup>

### **Professional Standards and Ethical Guidelines -**

**Moral Guidelines:** The ethical precepts of beneficence, justice, non-maleficence, and respect for autonomy should serve as the foundation for algorithms.<sup>42</sup>

Professional Training Professionals and social workers should be trained on the practical and ethical ramifications of algorithmic decision-making, as well as how to evaluate and question algorithmic results.<sup>43</sup>

## **CONCLUSION**

The integration of algorithmic tools into child welfare decision-making holds both promise and peril. While these systems can enhance consistency and efficiency, they also risk entrenching and amplifying existing inequalities if not carefully designed, implemented, and regulated. The evidence from Allegheny County and other jurisdictions underscores the urgent need for robust regulatory, ethical, and participatory frameworks to ensure that digital innovation serves the cause of justice, not discrimination.

Moving forward, policymakers, practitioners, and advocates must prioritise algorithmic accountability, procedural protections, ongoing oversight, and ethical integrity. Only by confronting the challenges of digital discrimination head-on can we build a child welfare system that is both effective and just in the digital age.

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<sup>41</sup> Hall SF and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>42</sup> Hall SF and others, 'A Systematic Review of Sophisticated Predictive and Prescriptive Analytics in Child Welfare: Accuracy, Equity, and Bias' (2023) Child and Adolescent Social Work Journal <https://par.nsf.gov/servlets/purl/10498097> accessed 18 June 2025.

<sup>43</sup> Elizabeth Kennedy, 'Algorithmic Harms in Child Welfare: Uncertainties in Practice, Organization, and Street-Level Decision-Making' (Montreal AI Ethics Institute, 2 March 2023) <https://montrealethics.ai/algorithmic-harms-in-child-welfare-uncertainties-in-practice-organization-and-street-level-decision-making/> accessed 14 June 2025.